



Beyond Routing Games: Network Formation Games



Network Games (NG)

- NG model the various ways in which selfish users (i.e., players) strategically interact in using a (either **communication, computer, social, etc.**) **network** (modelled as a **graph**)
- The **Internet routing game** is a particular type of **network congestion game**
- Other examples of NG: social network games, graphical games, **network design games**, **network creation games**, etc.
- Notice that each of these games is actually a **class of games**, where each element of the class is specified by the actual input graph, and it is called an **instance** of the game (i.e, it is a specific game)



Network Formation Games (NFG)

- NFG are NG that aim to capture two competing issues for players when **using** a network for **communication purposes**:
 - to minimize the **afforded cost**
 - to be provided with a high **quality of service**
- Two big categories of NFG:
 - **Network Design Games** (a.k.a. **Global Connection Games**): Users autonomously **design** a communication **subnetwork** embedded in an **already existing** network with the selfish goal of **sharing costs** in using it for a **point-to-point communication**
 - **Network Creation Games** (a.k.a. **Local Connection Games**): Users autonomously **form ex-novo** a network that connects them for **reciprocal communication** (e.g., downloading files in P2P networks, exchanging messages in social networks, etc.)



**First case study:
Network Design Games
(a.k.a. Global Connection
Games)**



Introduction

- Given a weighted graph G , a **Global Connection Game (GCG)** is a game that models the **selfish design** of a **communication subnetwork** of G , i.e., a set of **point-to-point communication paths**, where each path is associated with a player, and the selfish goal of each player is to **share the costs** for a joint use with other players of the edges on its selected path
- In other words, players:
 - pay for the links they personally use
 - benefit from sharing links with other players in the selected subnetwork



The formal definition of a GCG

- It is given a directed weighted graph $G=(V,E,c)$; c_e will denote the non-negative real weight of $e \in E$
- k players; each player is associated with a commodity (s_i, t_i) , with $s_i, t_i \in V$, and the strategy for a player i is to select a path P_i in G from s_i to t_i
- Let k_e denote the load of edge e , i.e., the number of players using e ; the cost of P_i for player i in a strategy profile $S=(P_1, \dots, P_k)$ is shared with all the other players using (part of) it, namely:

$$\text{cost}_i(S) = \sum_{e \in P_i} c_e / k_e$$

this cost-sharing scheme is called
fair or *Shapley cost-sharing mechanism*

The formal definition of a GCG (2)

- Given a strategy vector S , the **designed network** $N(S)$ is given by the union of all paths P_i
- Then, the **social-choice function** is the **utilitarian social cost**, namely the total cost of the designed network:

$$C(S) = \sum_i \text{cost}_i(S) = \sum_i \sum_{e \in P_i} c_e / k_e = \sum_{e \in N(S)} c_e$$

- Notice that each player has a favorable effect on the cost paid by other players (so-called *cross monotonicity*), as opposed to the *congestion model* of selfish routing



Open questions

- What is a **stable** network? We use NE as the solution concept, and we will seek for the existence of NE
- How to evaluate the **overall quality** of a stable network? We compare its cost to that of an **optimal** (in general, unstable) network, and we will try to estimate a bound on the efficiency loss resulting from selfishness
- Notice that the problem of finding an optimal network is a classic optimization problem (i.e., the **network design problem**), which is known to be NP-hard even if G is unweighted



Bounding the loss of efficiency

- Remind that a network is *optimal* or *socially efficient* if it minimizes the social cost (i.e., it minimizes the social-choice function)
- We know that the *PoA* is useful to estimate the loss of efficiency we may have in the *worst case*, as given by the ratio between the cost of a *worst* stable network and the cost of an optimal network
- But what about the ratio between the cost of a *best* stable network and the cost of an optimal network?

The price of stability (PoS)

■ **Definition** (Schulz & Moses, 2003): Given a (single-instance) game G and a **social-choice function** C (which depends on the payoff of **all** the players), let S be the set of all NE of G . If the payoff represents a **cost** (resp., a **utility**) for a player, let OPT be the outcome of G **minimizing** (resp., **maximizing**) C . Then, the **Price of Stability (PoS)** of G w.r.t. C is:

$$PoS_G(C) = \inf_{s \in S} \frac{C(s)}{C(OPT)} \left(\text{resp.}, \sup_{s \in S} \frac{C(s)}{C(OPT)} \right)$$

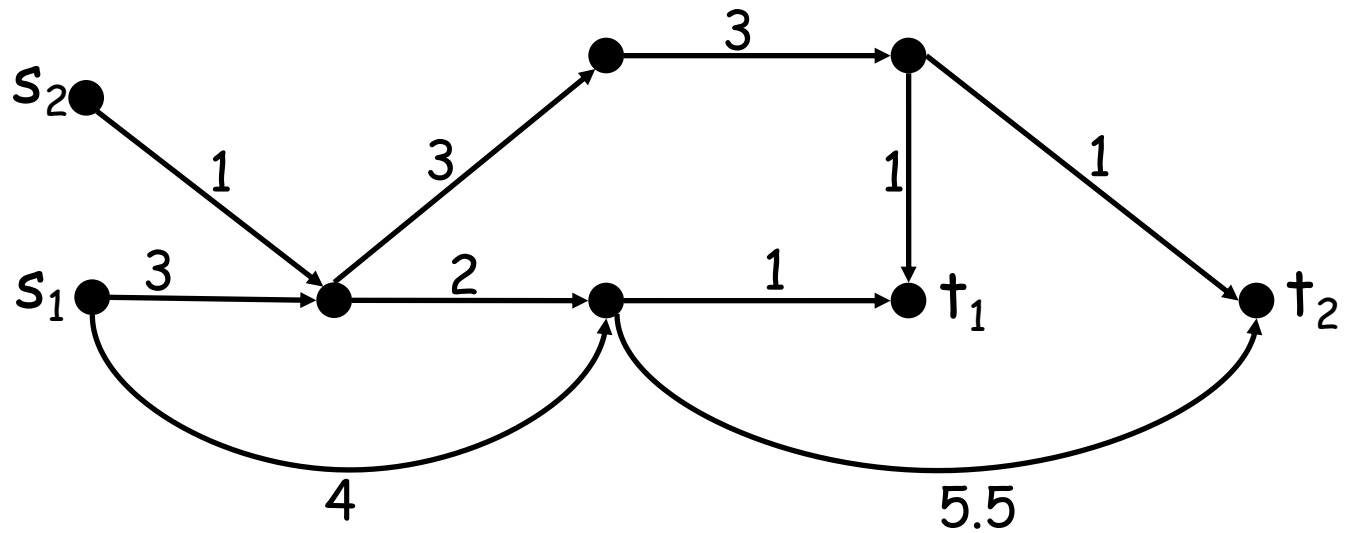
■ **Remark:** If G is a class of games (as for GCG), then its PoS is the **maximum/minimum** among the PoS of all the instances of G , depending on whether the payoff for a player is either a **cost** or a **utility**.



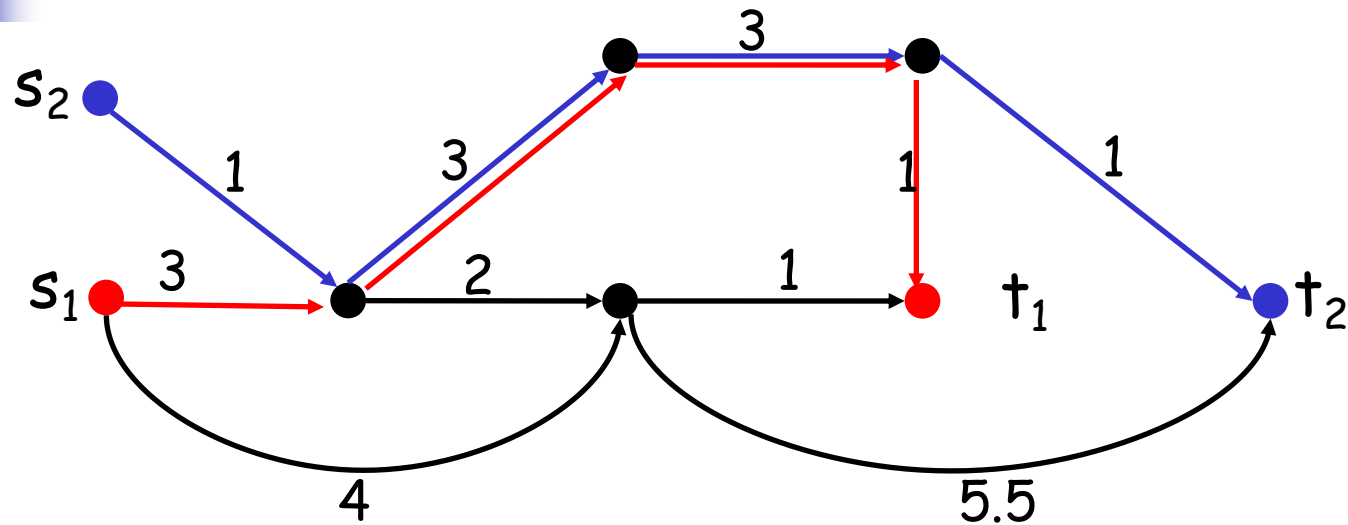
Some remarks

- PoA and PoS are (for positive s.c.f. C)
 - ≥ 1 for minimization (i.e., payoffs are **costs**) games
 - ≤ 1 for maximization (i.e., payoffs are **utilities**) games
- PoA and PoS are small when they are close to 1
- PoS is at least as close to 1 than PoA
- In a game with a **unique** NE, $PoA = PoS$
- Why studying the PoS?
 - sometimes a nontrivial bound is possible only for PoS
 - PoS quantifies a **lower bound** to the efficiency loss resulting from selfishness

An example



An example

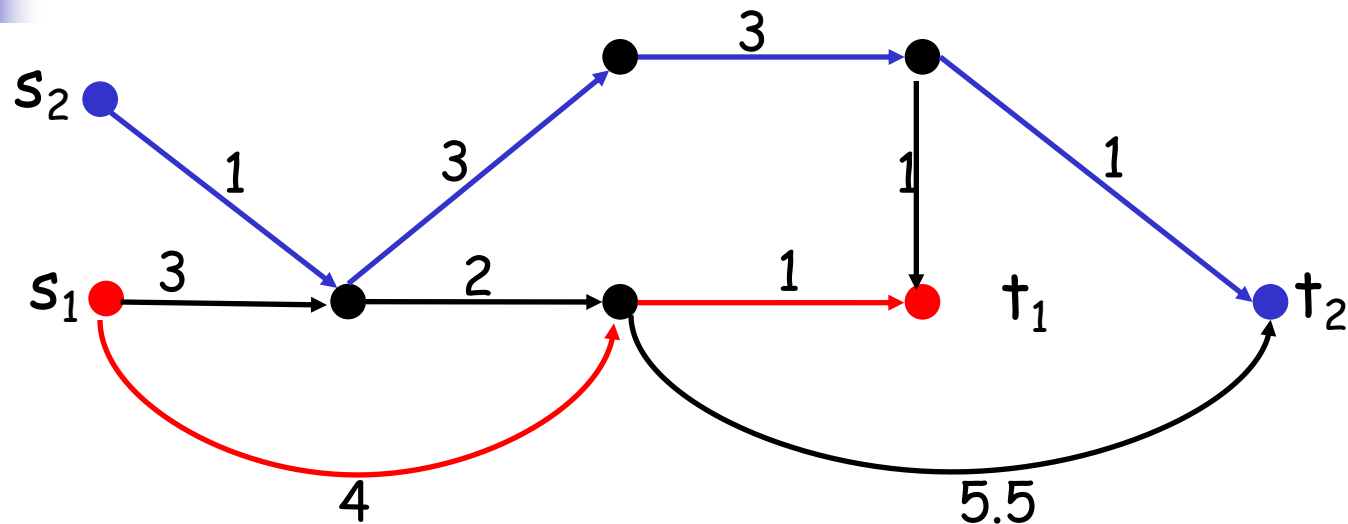


optimal network has cost 12

$$\text{cost}_1 = 7$$
$$\text{cost}_2 = 5$$

is it stable?

An example



...no!, player 1 can decrease its cost

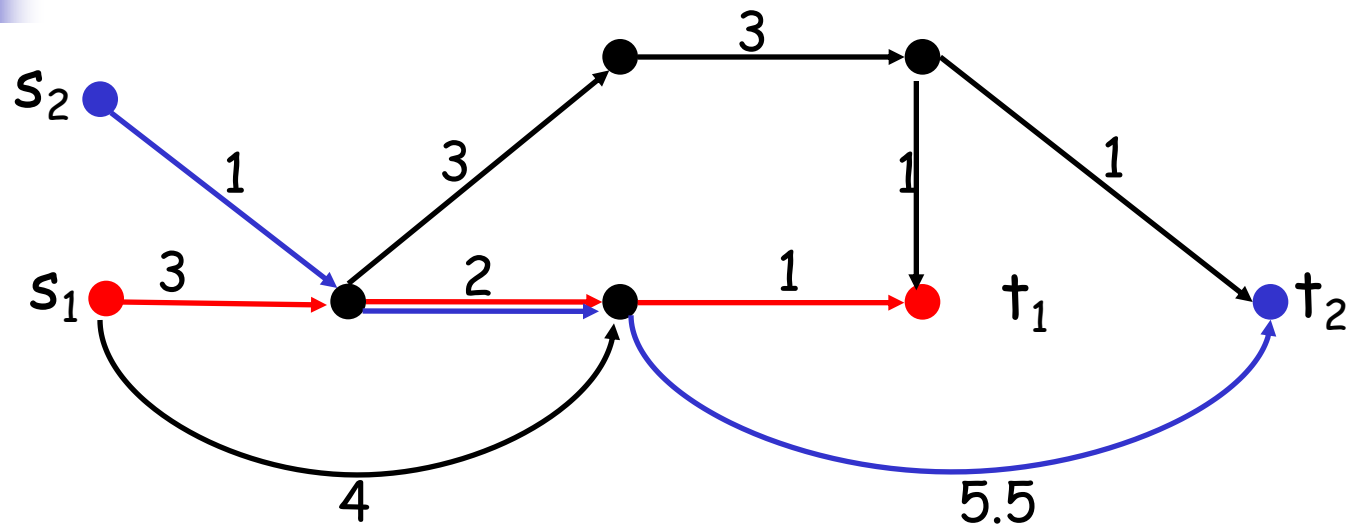
$$\text{cost}_1 = 5$$

$$\text{cost}_2 = 8$$

is it stable? ...yes, and has cost 13!

$$\Rightarrow \text{PoA} \geq 13/12, \text{PoS} \leq 13/12$$

An example



...a best possible NE:

$$\text{cost}_1 = 5$$

$$\text{cost}_2 = 7.5$$

the social cost is 12.5 \Rightarrow PoS = 12.5/12

Homework: find a worst possible NE

Theorem 1

Every instance of the GCG has a pure Nash equilibrium, and **best response dynamics** (i.e., that in which each player at each step selects its **best** available strategy) always converges.

Theorem 2

The PoA of a GCG with k players is at most k (i.e., **every instance** of the game has $\text{PoA} \leq k$), and this is **tight** (i.e., we can exhibit an **instance** of the game whose PoA is k).

Theorem 3

The PoS of a GCG with k players is at most H_k , the k -th harmonic number (i.e., **every instance** of the game has $\text{PoS} \leq H_k$), and this is **tight** (i.e., we can exhibit an **instance** of the game whose PoS is H_k)



The potential function method

For any *finite* game, an *exact potential function* Φ is a function that maps every strategy vector S to some real value and satisfies the following condition:

$\forall S=(s_1, \dots, s_i, \dots, s_k)$, let $s'_i \neq s_i$, and let $S'=(s_1, \dots, s'_i, \dots, s_k)$, then

$$\Phi(S) - \Phi(S') = \text{cost}_i(S) - \text{cost}_i(S').$$

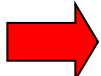
A (finite) game that does possess an exact potential function is called *potential game*

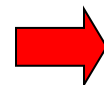
Lemma 1

Every potential game has at least one pure Nash equilibrium, namely the strategy vector \hat{S} that **minimizes** (resp., **maximizes**) Φ , assuming players' payoffs are costs (resp., utilities).

Proof (minimization): Observe that Φ is bounded. Then, starting from $\hat{S}=(\hat{s}_1,\dots,\hat{s}_i,\dots,\hat{s}_k)$, consider any move by a player i that results in a new strategy vector $S=(\hat{S}_{-i},s_i)=(\hat{s}_1,\dots,\hat{s}_{i-1},s_i,\dots,\hat{s}_k)$. Since $\Phi(\hat{S})$ is minimum, we have:

$$\underbrace{\Phi(\hat{S})-\Phi(S)}_{\leq 0} = \text{cost}_i(\hat{S})-\text{cost}_i(S)$$

 $\text{cost}_i(\hat{S}) \leq \text{cost}_i(S)$



player i cannot decrease its cost, thus \hat{S} is a NE.



Convergence in potential games

Observation: any state S with the property that $\Phi(S)$ cannot be decreased by changing any single component of S is a NE by the same argument. Furthermore, by definition, improving moves for players decrease the value of the potential function, which is **bounded**. Together, these observations imply the following result.

Lemma 2

In any finite potential game, best response dynamics always converges to a Nash equilibrium

⊗ However, it may be the case that converging to a NE takes an exponential (in the number of players) number of steps!



...turning our attention to the global connection game...

Let Ψ be the following function mapping any strategy vector S to a real value [Rosenthal 1973]:

$$\Psi(S) = \sum_{e \in N(S)} \Psi_e(S)$$

where (recall that k_e is the number of players using e)

$$\Psi_e(S) = c_e \cdot H_{k_e} = c_e \cdot (1 + 1/2 + \dots + 1/k_e).$$

Lemma 3 (Ψ is a potential function)

Let $S=(P_1,\dots,P_k)$, let P'_i be an alternative path for some player i , and define a new strategy vector $S'=(S_{-i},P'_i)$.

Then:

$$\Psi(S) - \Psi(S') = \text{cost}_i(S) - \text{cost}_i(S').$$

Proof:

When player i switches from P_i to P'_i , some edges of $N(S)$ increase their load by 1, some others decrease it by 1, and the remaining do not change it. Then, it suffices to notice that:

- If load of edge e increases by 1, its contribution to the potential function **increases** by $c_e/(k_e+1)$
- If load of edge e decreases by 1, its contribution to the potential function **decreases** by c_e/k_e

$$\Rightarrow \Psi(S) - \Psi(S') = \Psi(S) - \Psi(S - P_i + P'_i) = \Psi(S) -$$

$$(\Psi(S) - \sum_{e \in P_i} c_e/k_e + \sum_{e \in P'_i} c_e/(k_e+1)) = \text{cost}_i(S) - \text{cost}_i(S').$$





Existence of a NE

Theorem 1

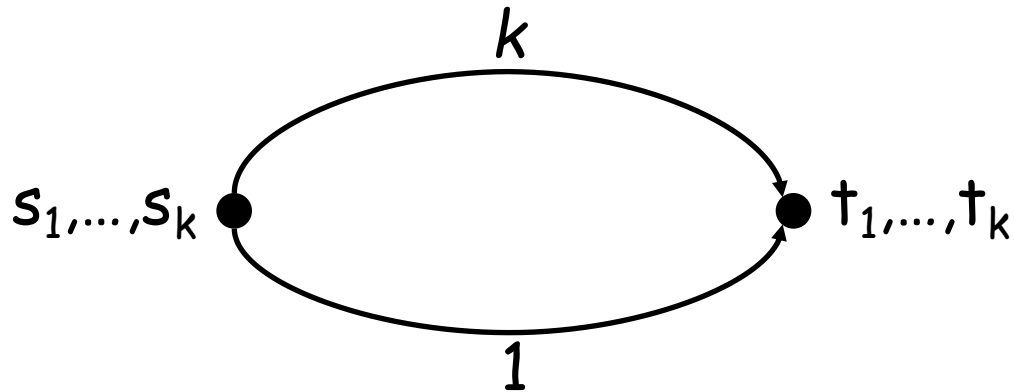
Every instance of the *GCG* has a pure Nash equilibrium, and best response dynamics always converges.

Proof: From Lemma 3, a *GCG* is a potential game, and from Lemma 1 and 2 best response dynamics converges to a pure NE. 

😊 It can be shown that finding a best response for a player is polynomial (it suffices to find a shortest path in G where each edges e is weighted as $c_e/(k_e+1)$)

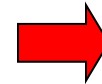
😞 Instead, it can be shown that finding a NE of cost at most C (and so, finding a best/worst NE) is NP-hard!

Price of Anarchy: a lower bound



optimal network has cost 1

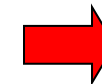
best NE: all players use the lower edge



PoS is 1



worst NE: all players use the upper edge



PoA is k



Upper-bounding the PoA

Theorem 2

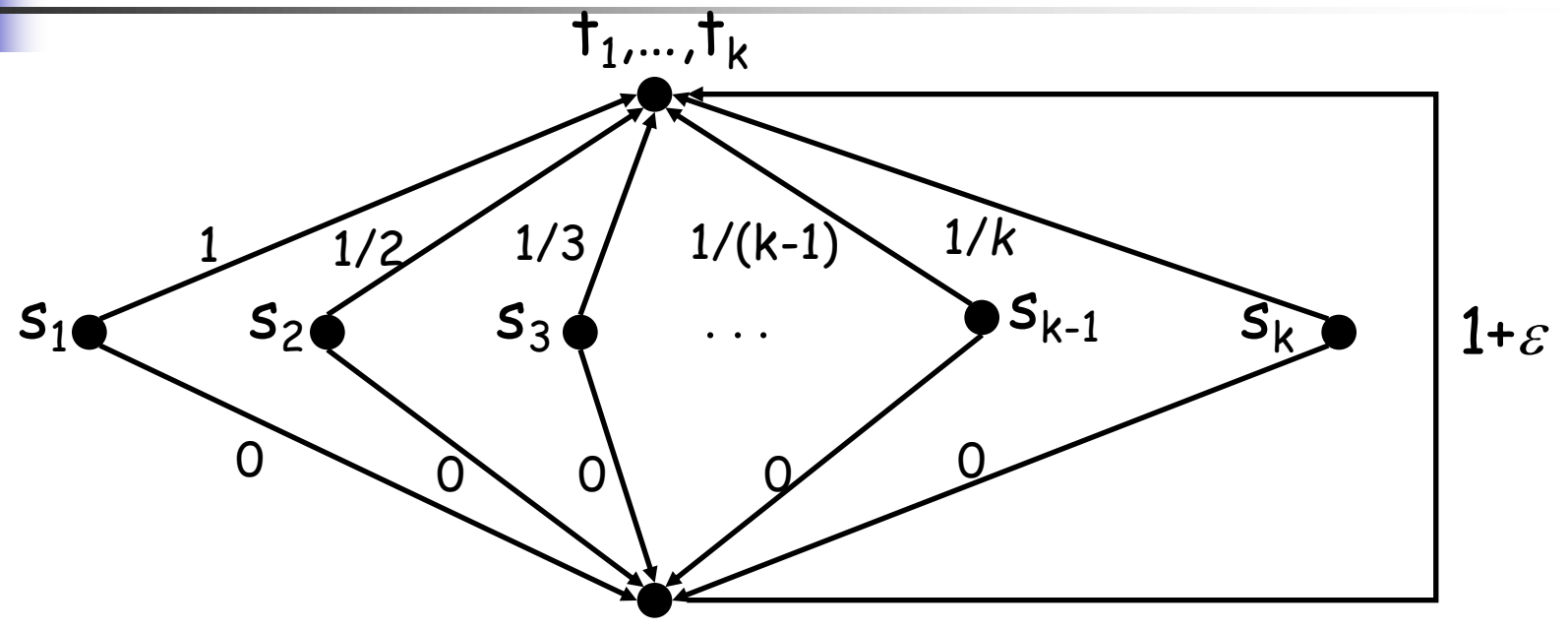
The price of anarchy in the global connection game with k players is at most k (and so, from the previous lower bound, this is tight).

Proof: Let $OPT=(P_1^*, \dots, P_k^*)$ denote the optimal set of paths (i.e., a set of paths minimizing C). Let Π_i be a **shortest path** in $G=(V, E, c)$ between s_i and t_i w.r.t. c , and let $c(\Pi_i) = \sum_{e \in \Pi_i} c_e$ be the length of such a path. Finally, let S be any NE. Observe that $cost_i(S) \leq c(\Pi_i)$ (otherwise the player i would change to Π_i). Then:

$$C(S) = \sum_{i=1}^k cost_i(S) \leq \sum_{i=1}^k c(\Pi_i) \leq \sum_{i=1}^k c(P_i^*) =$$
$$\sum_{i=1}^k \sum_{e \in P_i^*} c_e \leq \sum_{i=1}^k \sum_{e \in P_i^*} k \cdot c_e / k_e = \sum_{i=1}^k k \cdot cost_i(OPT) = k \cdot C(OPT). \quad \blacksquare$$

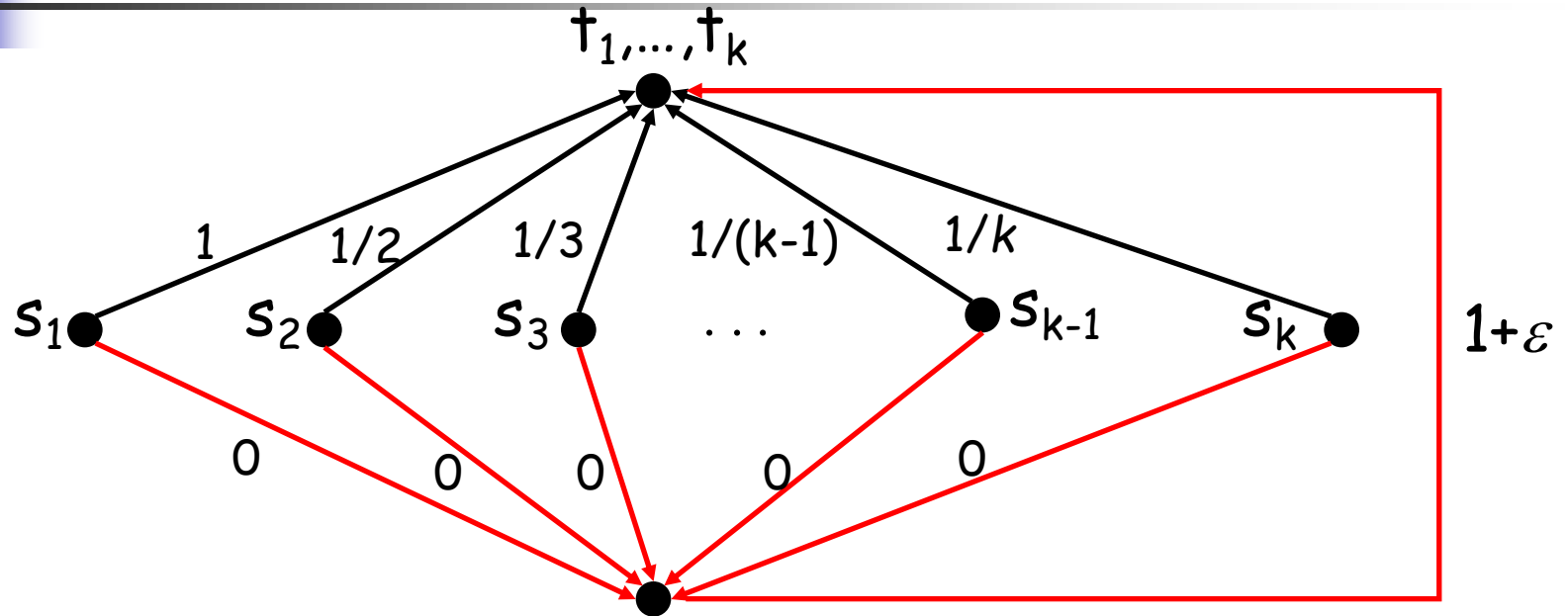
PoS for GCG: a lower bound

$\varepsilon \rightarrow 0$: small value



PoS for GCG: a lower bound

$\varepsilon \rightarrow 0$: small value

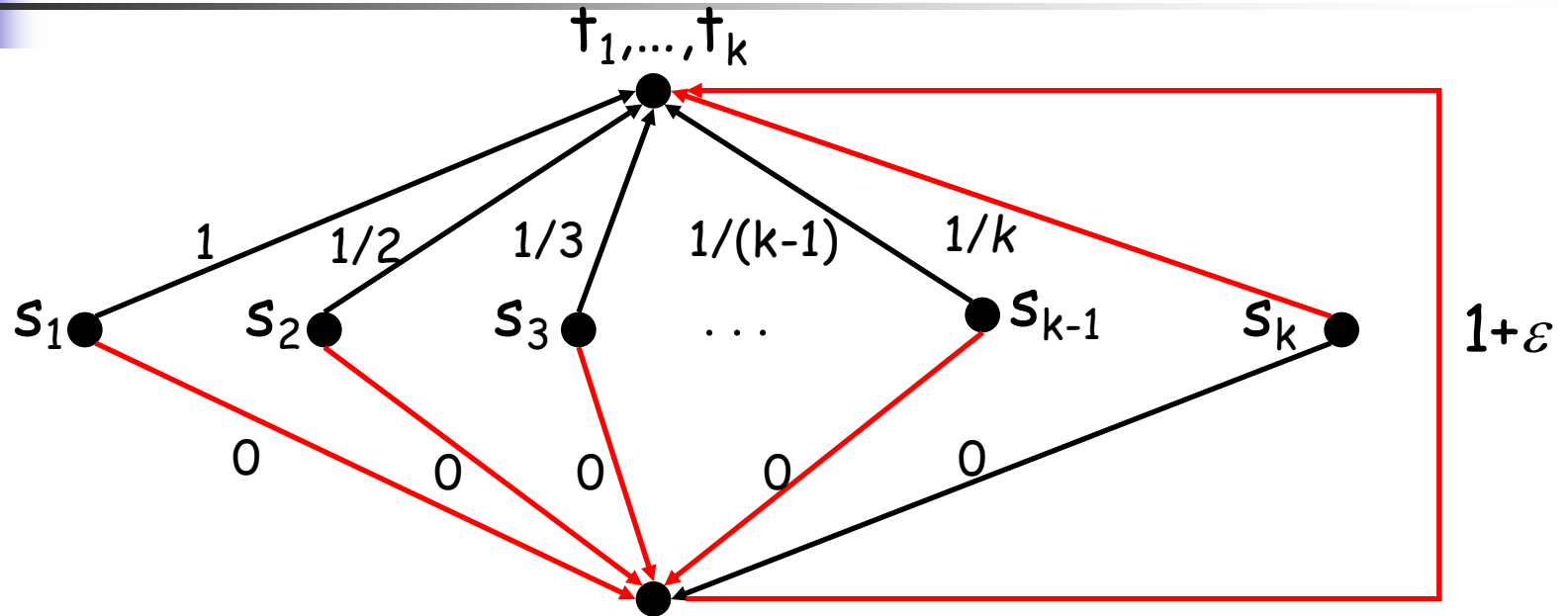


The optimal solution has a cost of $1 + \varepsilon$

is it stable?

PoS for GCG: a lower bound

$\varepsilon \rightarrow 0$: small value

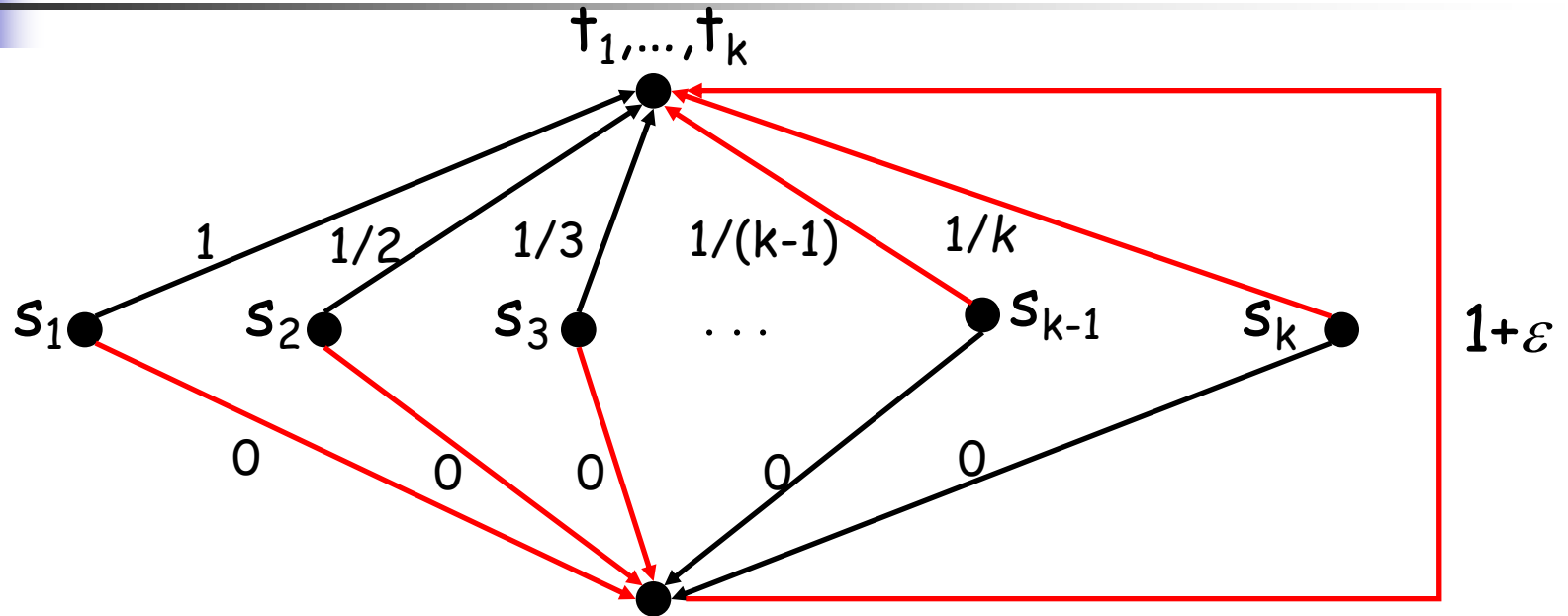


...no! player k can decrease its cost...

is it stable?

PoS for GCG: a lower bound

$\varepsilon \rightarrow 0$: small value

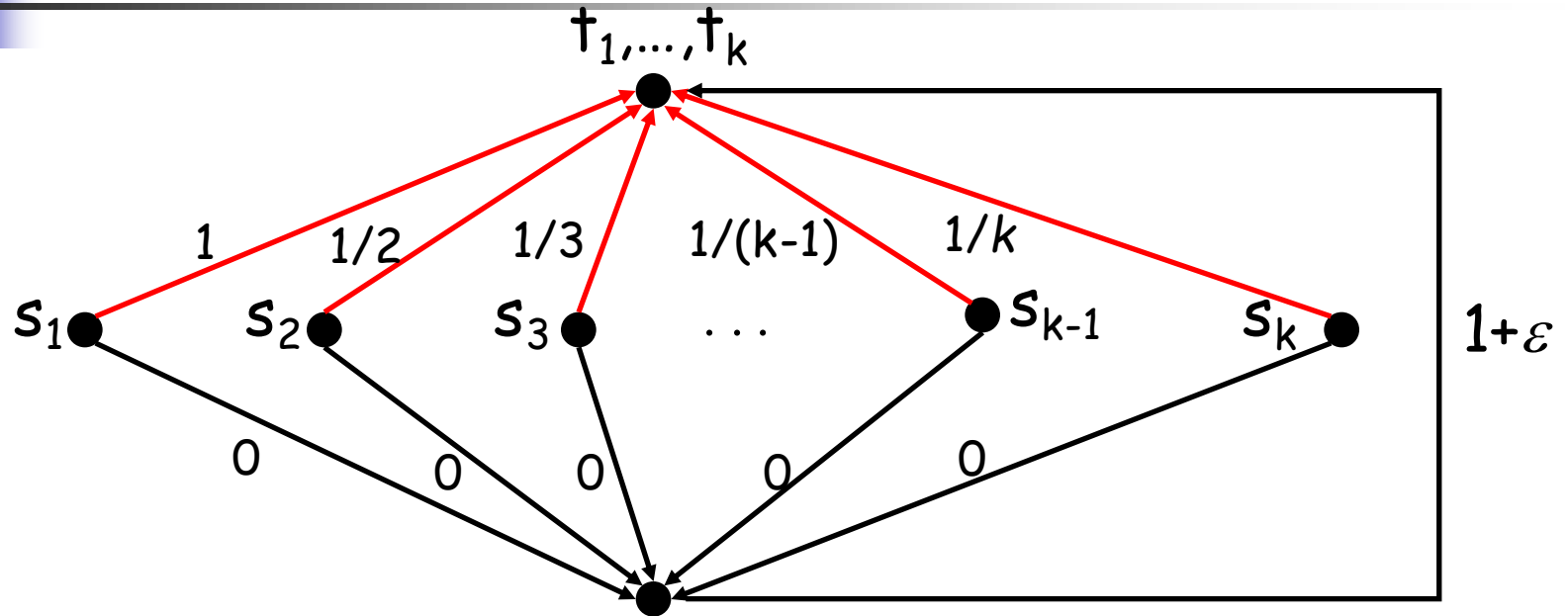


...no! player $k-1$ can decrease its cost...

is it stable?

PoS for GCG: a lower bound

$\varepsilon \rightarrow 0$: small value



The only stable network

social cost: $C(S) = \sum_{j=1}^k 1/j = H_k \leq \ln k + 1$ k -th *harmonic number*

Lemma 4

Suppose that we have a potential game with potential function Φ , and assume that for any outcome S we have

$$C(S)/A \leq \Phi(S) \leq B C(S)$$

for some $A, B > 0$. Then the price of stability is at most AB .

Proof:

Let \hat{S} be the strategy vector minimizing Φ (i.e., \hat{S} is a NE, from Lemma 1). Let S^* be the strategy vector minimizing the social cost

we have:

$$C(\hat{S})/A \leq \Phi(\hat{S}) \leq \Phi(S^*) \leq B C(S^*)$$

$$\Rightarrow \text{PoS} \leq C(\hat{S})/C(S^*) \leq A \cdot B.$$



Lemma 5 (Bounding Ψ)

For any strategy vector S in the GCG , we have:

$$C(S) \leq \Psi(S) \leq H_k C(S).$$

Proof: Indeed:

$$\Psi(S) = \sum_{e \in N(S)} \Psi_e(S) = \sum_{e \in N(S)} c_e \cdot H_{k_e}$$

$$\Rightarrow \Psi(S) \geq C(S) = \sum_{e \in N(S)} c_e$$

$$\text{and } \Psi(S) \leq H_k \cdot C(S) = \sum_{e \in N(S)} c_e \cdot H_k.$$





Upper-bounding the PoS

Theorem 3

The price of stability in the global connection game with k players is at most H_k , the k -th harmonic number (and so, from the previous lower bound, this is tight).

Proof: From Lemma 3, a GCG is a potential game, and from Lemma 5 and Lemma 4 (with $A=1$ and $B=H_k$), its PoS is at most H_k .

